# Employing asremlPlus, in conjunction with asreml, to calculate and use information criteria

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This vignette illustrates the facilities in asremlPlus (Brien, 2023), in conjunction with asreml (Butler et al., 2023), for calculating and using information. Here, asremlPlus and asreml are packages for the R Statistical Computing environment (R Core Team, 2023).

It is divided into the following main sections:

- 1. Set up the maximal model for this experiment
- 2. Obtaining information criteria for separate models
- 3. Obtaining information criteria for a prescribed sequence of model changes
- 4. Using information criteria to decide model changes

### 1. Set up the maximal model for this experiment

```
library(knitr)
opts_chunk$set("tidy" = FALSE, comment = NA)
suppressMessages(library(asreml, quietly=TRUE))

## Offline License checked out Sun Nov 5 13:28:11 2023
packageVersion("asreml")

## [1] '4.2.0.276'
suppressMessages(library(asremlPlus))
packageVersion("asremlPlus")

## [1] '4.4.20'
options(width = 100)
```

#### Get data available in asremlPlus

The data are from a 1976 spring wheat experiment and are taken from Gilmour et al. (1995). An analysis is presented in the asrem1 manual by Butler et al. (2023, Section 7.6), although they suggest that it is a barley experiment.

```
data(Wheat.dat)
```

#### Fit the maximal model

In the following a model is fitted that has the terms that would be included for a balanced lattice. In addition, a term WithinColPairs has been included to allow for extraneous variation arising between pairs of adjacent

lanes. Also, separable ar1 residual autocorrelation has been included. This model represents the maximal anticipated model,

Warning in asreml(yield  $\sim$  WithinColPairs + Variety, random =  $\sim$ Rep/(Row + : Some components changed by more than 1% on the last iteration

The warning from asreml is probably due to a bound term.

#### Initialize a testing sequence by loading the current fit into an asrtests object

```
max.asrt <- as.asrtests(max.asr, NULL, NULL)</pre>
```

#### Check for and remove any boundary terms

```
max.asrt <- rmboundary(max.asrt)</pre>
summary(max.asrt$asreml.obj)$varcomp
                                                   z.ratio bound %ch
                         component
                                      std.error
Rep:Row
                                                               P 0.0
                      4.293282e+03 3.199458e+03 1.3418779
Rep:Column
                                                               P 0.7
                      1.575689e+02 1.480357e+03 0.1064398
units
                      5.742689e+03 1.652457e+03 3.4752438
                                                               P 0.0
Row:Column!R
                      4.706787e+04 2.515832e+04 1.8708669
                                                               P 0.0
Row:Column!Row!cor
                      7.920301e-01 1.014691e-01 7.8056280
                                                               U 0.0
Row:Column!Column!cor 8.799559e-01 7.370402e-02 11.9390486
                                                               U 0.0
print(max.asrt, which = "testsummary")
```

#### Sequence of model investigations

(If a row has NA for p but not denDF, DF and denDF relate to fixed and variance parameter numbers)

```
terms DF denDF p AIC BIC action
1 Rep 1 NA NA NA NA Boundary
```

Rep has been removed because it has been constrained to zero. Following the recommendation of Littel et al. (2006, p. 150), the bound on all variance components is set to unconstrained (U) using setvariances.asreml so as to avoid bias in the estimate of the residual variance. Alternatively, one could move Rep to the fixed model.

#### Unbind Rep, Row and Column components and reload into an asrtests object

Warning in asreml(fixed = yield  $\sim$  WithinColPairs + Variety, random =  $\sim$ Rep/(Row + : Some components changed by more than 1% on the last iteration

```
max.asrt <- as.asrtests(max.asr, NULL, NULL)</pre>
max.asrt <- rmboundary(max.asrt)</pre>
summary(max.asrt$asreml.obj)$varcomp
                                                      z.ratio bound %ch
                           component
                                         std.error
Rep
                       -2458.3485841 1.197491e+03 -2.0529167
                                                                   U 0.0
Rep:Row
                        5008.7151486 3.401335e+03 1.4725732
                                                                   U 0.0
Rep:Column
                         916.4641198 1.699576e+03
                                                    0.5392309
                                                                   U 0.2
                                                                   P 0.0
units
                        5959.0220817 1.609649e+03 3.7020634
Row: Column! R
                       46637.6303429 2.724392e+04 1.7118545
                                                                   P 0.0
Row:Column!Row!cor
                           0.8150590 1.000281e-01 8.1483012
                                                                   U 0.0
```

0.8856824 7.492514e-02 11.8208968

```
print(max.asrt, which = "testsummary")
```

Row: Column! Column! cor

#### Sequence of model investigations

(If a row has NA for p but not denDF, DF and denDF relate to fixed and variance parameter numbers)

U 0.0

```
[1] terms DF denDF p AIC BIC action <0 rows> (or 0-length row.names)
```

Now the Rep component estimate is negative.

The test.summary output shows that no changes have been made to the model loaded using as.asrtests. The pseudo-anova table shows that Varieties are highly significant (p < 0.001)

### 2. Obtaining information criteria for separate models

0 1647.191 1746.542 -790.5957

The method infoCriteria has two methods for calculating information criteria. One, infoCriteria.asreml, is a method for asreml objects and the other, infoCriteria.list, if for 'listobjects, the components of thelistbeingareml' objects.

#### Single models

Firstly, infoCriteria is called with the default IClikelihood, which is REML. Then it is called with IClikelihood set to full (Verbyla, 2019).

```
infoCriteria(max.asr)

fixedDF varDF NBound   AIC   BIC loglik
1     0     7     0 1396.34 1416.082 -691.17

infoCriteria(max.asr, IClikelihood = "full")

Warning in asreml(fixed = yield ~ WithinColPairs + Variety, random = ~Rep/(Row + : Log-likelihood not converged
   fixedDF varDF NBound   AIC   BIC loglik
```

#### A list of models

26

1

Now, a second model, from which the withinColPairs term has been omitted, is fitted; to be consistent, the variance components are unconstrained using setvariances.asreml. Then the asreml objects for this

model and the maximal model are combined into a list and a data.frame produced that includes their information criteria.

Warning in asreml(yield  $\sim$  Variety, random =  $\sim$ Rep/(Row + Column) + units, : Some components changed by more than 1% on the last iteration

Warning in asreml(fixed = yield  $\sim$  Variety, random =  $\sim$ Rep/(Row + Column) + : Some components changed by more than 1% on the last iteration

```
mods <- list(max = max.asr, m1 = m1.asr)
ic <- infoCriteria(mods, IClikelihood = "full")
print(ic)</pre>
```

```
    fixedDF
    varDF
    NBound
    AIC
    BIC
    loglik

    max
    26
    7
    0 1647.191
    1746.542
    -790.5957

    m1
    25
    7
    0 1645.318
    1741.658
    -790.6588
```

# 3. Obtaining information criteria for a prescribed sequence of model changes

The use of changeTerms.asrtests is demonstrated for a sequence of models, starting with the maximal model.

#### Drop the term for within Column pairs (a post hoc factor)

Warning in asreml(fixed = yield ~ WithinColPairs + Variety, random = ~Rep/(Row + : Log-likelihood not converged

Warning in asreml(fixed = yield  $\sim$  Variety, random =  $\sim$ Rep + units + Rep:Row + : Some components changed by more than 1% on the last iteration

Warning in asreml(fixed = yield ~ Variety, random = ~Rep + units + Rep:Row + : Some components changed by more than 1% on the last iteration

```
print(current.asrt, which = "testsummary", omit.columns = "p")
```

#### Sequence of model investigations

(If a row has NA for p but not denDF, DF and denDF relate to fixed and variance parameter numbers)

```
terms DF denDF AIC BIC action
Maximal model 26 7 1647.191 1746.542 Starting model
Drop withinColPairs 25 7 1645.325 1741.666 Changed fixed
```

So the same values of the information criteria have been obtained as when infoCriteria.list was used on a list containing the asreml objects for the two models. The differences is that here there is ultimately only one fitted model, the model stored in the asreml object in the asrtests object named current.asrt: this is the model with withinColPairs omitted.

Note this use of the omit.columns argument from print.test.summary to omit the irrelevant column p from the test.summary.

#### Drop nugget term

Warning in asreml(fixed = yield ~ Variety, random = ~Rep + Rep:Row + Rep:Column, : Some components changed by more than 1% on the last iteration

Warning in asreml(fixed = yield  $\sim$  Variety, random =  $\sim$ Rep + Rep:Row + Rep:Column, : Some components changed by more than 1% on the last iteration

#### Check Row autocorrelation

#### Sequence of model investigations

(If a row has NA for p but not denDF, DF and denDF relate to fixed and variance parameter numbers)

```
terms DF denDF AIC BIC action
1 Maximal model 26 7 1647.191 1746.542 Starting model
2 Drop withinColPairs 25 7 1645.325 1741.666 Changed fixed
3 Drop units 25 6 1650.126 1743.456 Changed random
4 Row autocorrelation 25 5 1660.882 1751.201 Changed residual
```

## 4. Using information criteria to decide model changes

This sections illustrates the use of changeModelOnIC.asrtests to decide between consecutive models in a sequence of models. The default information criterion to use for this is the AIC. However, which.IC can be used to specify the use of the BIC or both. Here we use the AIC and the full likelihood.

#### Check the term for within Column pairs (a post hoc factor)

As before, we start with the maximal model, in which the variance components have been unconstrained and look to decide whether of not to drop the withinColPairs term.

Warning in asreml(fixed = yield ~ WithinColPairs + Variety, random = ~Rep/(Row + : Log-likelihood not converged

Warning in asreml(fixed = yield ~ Variety, random = ~Rep + units + Rep:Row + : Some components changed by more than 1% on the last iteration

Warning in asreml(fixed = yield ~ Variety, random = ~Rep + units + Rep:Row + : Some components changed by more than 1% on the last iteration

```
print(current.asrt, which = "testsummary", omit.columns = "p")
```

#### Sequence of model investigations

(If a row has NA for p but not denDF, DF and denDF relate to fixed and variance parameter numbers)

```
terms DF denDF AIC BIC action
Maximal model 26 7 1647.191446 1746.542411 Starting model
withinColPairs -1 0 -1.866103 -4.876738 Swapped
```

Given the warning about a lack of convergence, we use iterate.asrtests to perform additional iterations of the fitting process. It seems that it was successful.

It can be seen from the test.summary that the term has been swapped out and this has the effect of reducing the number of fixed parameters by one and makes no change to the variance parameters.

#### Check the nugget term

Warning in asreml(fixed = yield ~ Variety, random = ~Rep + Rep:Row + Rep:Column, : Some components changed by more than 1% on the last iteration

Warning in asreml(fixed = yield  $\sim$  Variety, random =  $\sim$ Rep + Rep:Row + Rep:Column, : Some components changed by more than 1% on the last iteration

#### Check Row autocorrelation

Warning in asreml(fixed = yield ~ Variety, random = ~Rep + units + Rep:Row + : Log-likelihood not converged

Warning in asreml(fixed = yield  $\sim$  Variety, random =  $\sim$ Rep + units + Rep:Row + : Some components changed by more than 1% on the last iteration

```
Warning in asreml(fixed = yield ~ Variety, random = ~Rep + units + Rep:Row + : Log-likelihood not
converged
Warning in asreml(fixed = yield ~ Variety, random = ~Rep + units + Rep:Row + : Some components
changed by more than 1% on the last iteration
Warning in newfit.asreml(asreml.obj, fixed. = fix.form, random. = ran.form, :
```

# Check Column autocorrelation (depends on whether Row autocorrelation retained)

Warning in infoCriteria.asreml(asreml.obj, IClikelihood = ic.lik, bound.exclusions = bound.exclusions):
Row:Column!Row!cor

Warning in rmboundary.asrtests(as.asrtests(asreml.obj, wald.tab, test.summary, : In analysing yield, es Row:Column!Row!cor

Warning in infoCriteria.asreml(new.asrtests.obj\$asreml.obj, IClikelihood = ic.lik, : The following bound Row:Column!Row!cor

#### Output the results

```
print(current.asrt, which = "test", omit.columns = "p")
```

#### Sequence of model investigations

(If a row has NA for p but not denDF, DF and denDF relate to fixed and variance parameter numbers)

```
terms DF denDF
                                        ATC
                                                                              action
       Maximal model 26
                        7 1.647191e+03 1.746542e+03
1
                                                                      Starting model
2
      withinColPairs -1
                            0 -1.866103e+00 -4.876738e+00
                                                                             Swapped
                           -1 4.801053e+00 1.790418e+00
               units 0
                                                                           Unswapped
4 Row autocorrelation 0
                            0 -7.342295e-03 -7.342295e-03 Unchanged - new unconverged
5 Col autocorrelation 0
                           -2 1.947985e+01 1.345858e+01
                                                                           Unswapped
summary(current.asrt$asreml.obj)$varcomp
```

```
component std.error z.ratio bound %ch
Rep -2391.8946799 1.194671e+03 -2.002136 U 0.4
Rep:Row 5035.4828349 3.406065e+03 1.478387 U 0.3
```

```
Rep:Column
                        761.9005140 1.612048e+03 0.472629
                                                                U 1.2
units
                                                                P 0.1
                       5933.1408473 1.610819e+03
                                                  3.683306
Row:Column!R
                      45970.2439168 2.635029e+04 1.744582
                                                                P 0.0
Row:Column!Row!cor
                          0.8101593 9.995689e-02 8.105087
                                                                U 0.1
Row:Column!Column!cor
                          0.8846965 7.503099e-02 11.791081
                                                                U 0.0
```

The test.summary shows us that the model without the autocorrelation failed to converge and so no change was made to the model. It, and the messages from checking the Column autocorrelation, also show us that the omission of the Column autocorrelation resulted in the Row autocorrelation becoming bound. That is, dropping the Column autocorrelation resulted in the dropping of two variance parameters

The function printFormulae.asreml is used to display the fitted model.

```
printFormulae(current.asrt$asreml.obj)
```

#### #### Formulae from asreml object

```
fixed: yield ~ Variety
```

random: ~ Rep + units + Rep:Row + Rep:Column

residual: ~ ar1(Row):ar1(Column)

#### References

Brien, C. J. (2023) asremlPlus: Augments ASReml-R in fitting mixed models and packages generally in exploring prediction differences. Version 4.4.20. https://cran.r-project.org/package=asremlPlus/ or http://chris.brien.name/rpackages/.

Butler, D. G., Cullis, B. R., Gilmour, A. R., Gogel, B. J. and Thompson, R. (2023). ASReml-R Reference Manual Version 4.2. VSN International Ltd, https://asreml.kb.vsni.co.uk/.

Gilmour, A. R., Thompson, R., & Cullis, B. R. (1995). Average Information REML: An Efficient Algorithm for Variance Parameter Estimation in Linear Mixed Models. *Biometrics*, **51**, 1440–1450.

Littell, R. C., Milliken, G. A., Stroup, W. W., Wolfinger, R. D., & Schabenberger, O. (2006). SAS for Mixed Models (2nd ed.). Cary, N.C.: SAS Press.

R Core Team (2023) R: A language and environment for statistical computing. Vienna, Austria: R Foundation for Statistical Computing. https://www.r-project.org/.

Verbyla, A. P. (2019). A note on model selection using information criteria for general linear models estimated using REML. Australian & New Zealand Journal of Statistics, 61, 39-50. https://doi.org/10.1111/anzs.12254/.